

Practitioner's Docket No. 740789-052110

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CHAPTER II

**TO THE UNITED STATES ELECTED OFFICE (EO/US)  
(ENTRY INTO U.S. NATIONAL PHASE UNDER CHAPTER II)**

INTERNATIONAL APPLICATION NO.	INTERNATIONAL FILING DATE	PRIORITY DATE CLAIMED
PCT/GB00/01675 ✓	02 May 2000 (02.05.00) ✓	01 May 1999 (01.05.99)

TITLE OF INVENTION

METHOD OF ANALYSIS OF MEDICAL SIGNALS

APPLICANTS

ADDISON, Paul, Stanley and WATSON, James, Nicholas ✓

Box PCT

Assistant Commissioner for Patents  
Washington D.C. 20231

ATTENTION: EO/US

## CERTIFICATE OF MAILING

I hereby certify that this correspondence, on the date shown below, is being deposited with the United States Postal Service with sufficient postage as Express Mail Label No. EL565093291US in an envelope addressed to Box PCT, Assistant Commissioner of Patents, Washington, D.C. 20231, Attention: EO/US.

Date: 01 November 2001
  
Nicole M. Gignac

1. Applicant herewith submits to the United States Elected Office (EO/US) the following items under 35 U.S.C. 371:
  - a. [X] This express request to immediately begin national examination procedures (35 U.S.C. Section 371(f)).
  - b. [X] The U.S. National Fee (35 U.S.C. Section 371(c)(1)) and other fees (37 C.F.R. Section 1.492) as indicated below:

In re application of: ADDISON, P.S., et al. Group: Not yet assigned  
Application No.: Not Yet Assigned Examiner: Not yet assigned  
(National Phase Entry of PCT/GB00/01675)  
Filed: Herewith  
For: METHOD OF ANALYSIS OF MEDICAL SIGNALS

2. Fees

CLAIMS FEE	(1) FOR	(2) NUMBER FILED	(3) NUMBER EXTRA	(4) RATE	(5) CALCULA- TIONS
[ ]*	TOTAL CLAIMS	- 20 =	0	x \$ 18.00 =	\$ 0
	INDEPENDENT CLAIMS	- 3 =	0	x \$ 84.00 =	\$ 0
	MULTIPLE DEPENDENT CLAIM(S) (if applicable) + \$280.00				\$ 0
BASIC FEE**	<p>[ ] U.S. PTO WAS INTERNATIONAL PRELIMINARY EXAMINATION AUTHORITY Where an international preliminary examination fee as set forth in Section 1.482 has been paid on the international application to the U.S. PTO: [ ] and the international preliminary examination report states that the criteria of novelty, inventive step (non-obviousness) and industrial activity, as defined in PCT Article 33(2) to (4) have been satisfied for all the claims presented in the application entering the national stage (37 C.F.R. Section 1.492(a)(4)) ..... \$100.00 [ ] and the above requirements are not met (37 C.F.R. Section 1.492(a)(1)) ..... \$710.00</p> <p>[X] U.S. PTO WAS NOT INTERNATIONAL PRELIMINARY EXAMINATION AUTHORITY Where no international preliminary examination fee as set forth in Section 1.482 has been paid to the U.S. PTO, and payment of an international search fee as set forth in Section 1.445(a)(2) to the U.S. PTO: [ ] has been paid (37 C.F.R. 1.492(a)(2)) ..... \$740.00 [ ] has not been paid (37 C.F.R. 1.492(a)(3)) ..... \$1,040.00 [X] where a search report on the international application has been prepared by the European Patent Office or the Japanese Patent Office (37 C.F.R. Section 1.492(a)(5)) ..... \$890.00</p>				\$890.00
	Total of above Calculations				= \$ 890.00
SMALL ENTITY	Reduction by 1/2 for filing by small entity, if applicable. Applicant asserts small entity status.				- \$ 445.00
	Subtotal				\$ 445.00
	Total National Fee				\$ 445.00
	Fee for recording the enclosed assignment document \$40.00 (37 C.F.R. 1.21(h)). (See Item 13 below). See attached "ASSIGNMENT COVER SHEET".				0.00
TOTAL	Total Fees enclosed				\$ 445.00

- i. [X] A check in the amount of \$ 445.00 to cover the above fees is enclosed.  
ii. [X] Please charge Account No. 50-0850 - 37 C.F.R. Section 1.492(b), (c) and (d) (presentation of extra claims)  
A duplicate copy of this sheet is enclosed.

In re application of: GILCHRIST, Thomas  
Application No.: Not Yet Assigned  
(National Phase Entry of PCT/GB99/03077)  
Filed: Herewith  
For: BONE REPAIR COMPOSITE MATERIAL

Group: Not yet assigned  
Examiner: Not yet assigned

3. ☒ A copy of the International application as filed (35 U.S.C. Section 371(c)(2)):
- a. ☐ is transmitted herewith.
  - b. ☐ is not required, as the application was filed with the United States Receiving Office.
  - c. ☒ has been transmitted
    - i. ☒ by the International Bureau.  
Date of mailing of the application: \_\_\_\_\_
    - ii. ☐ by applicant on \_\_\_\_\_  
Date \_\_\_\_\_
4. ☒ A translation of the International application into the English language (35 U.S.C. Section 371(c)(2)):
- a. ☐ is transmitted herewith.
  - b. ☒ is not required as the application was filed in English.
  - c. ☐ was previously transmitted by applicant on \_\_\_\_\_  
Date \_\_\_\_\_
  - d. ☐ will follow.
5. ☐ Amendments to the claims of the International application under PCT Article 19 (35 U.S.C. Section 371(c)(3)):
- a. ☐ are transmitted herewith.
  - b. ☐ have been transmitted
    - i. ☐ by the International Bureau.  
Date of mailing of the amendment (from form PCT/IB/308): \_\_\_\_\_
    - ii. ☐ by applicant on \_\_\_\_\_  
Date \_\_\_\_\_
  - c. ☐ have not been transmitted as
    - i. ☐ applicant chose not to make amendments under PCT Article 19.  
Date of mailing of Search Report \_\_\_\_\_
    - ii. ☐ the time limit for the submission of amendments has not yet expired.  
The amendments or a statement that amendments have not been made will be transmitted before the expiration of the time limit under PCT Rule 46.1.
6. ☐ A translation of the amendments to the claims under PCT Article 19 (38 U.S.C. Section 371(c)(3)):
- a. ☐ is transmitted herewith.
  - b. ☐ is not required as the amendments were made in the English language.
  - c. ☐ has not been transmitted for reasons indicated at point 5(c) above.

In re application of: ADDISON, P.S., et al. Group: Not yet assigned  
Application No.: Not Yet Assigned Examiner: Not yet assigned  
(National Phase Entry of PCT/GB00/01675)  
Filed: Herewith  
For: METHOD OF ANALYSIS OF MEDICAL SIGNALS

7. ☐ A copy of the international examination report (PCT/IPEA/409)  
☐ is transmitted herewith.  
☐ is not required as the application was filed with the United States Receiving Office.
8. ☐ Annex(es) to the international preliminary examination report  
a. ☐ is/are transmitted herewith.  
\*Applicants request entry of the annex to the IPER to the International Application if such has not been done.  
b. ☐ is/are not required as the application was filed with the United States Receiving Office.  
c. ☐ there are no annexes.
9. ☐ A translation of the annexes to the international preliminary examination report  
a. ☐ is transmitted herewith.  
b. ☐ is not required as the annexes are in the English language.  
c. ☐ there are no annexes.
10. ☒ An oath or declaration of the inventor (35 U.S.C. Section 371(c)(4)) complying with 35 U.S.C. 115  
a. ☐ was previously submitted by applicant on \_\_\_\_\_  
Date  
b. ☒ is submitted herewith, and such oath or declaration  
i. ☒ is attached to the application.  
ii. ☐ identifies the application and any amendments under PCT Article 19 that were transmitted as stated in points 3(b) or 3(c) and 5(b); and states that they were reviewed by the inventor as required by 37 C.F.R. Section 1.70.  
c. ☐ will follow.

Other document(s) or information included:

11. ☒ An International Search Report (PCT/ISA/210) or Declaration under PCT Article 17(2)(a):  
a. ☒ is transmitted herewith.  
b. ☐ has been transmitted by the International Bureau.  
Date of mailing (from form PCT/IB/308): \_\_\_\_\_  
c. ☐ is not required, as the application was searched by the United States International Searching Authority.  
d. ☐ will be transmitted promptly upon request.  
e. ☐ has been submitted by applicant on \_\_\_\_\_  
Date

In re application of: GILCHRIST, Thomas  
Application No.: Not Yet Assigned  
(National Phase Entry of PCT/GB99/03077)  
Filed: Herewith  
For: BONE REPAIR COMPOSITE MATERIAL

Group: Not yet assigned  
Examiner: Not yet assigned

12. ☒ An Information Disclosure Statement under 37 C.F.R. Sections 1.97 and 1.98:  
a. ☐ is transmitted herewith.  
Also transmitted herewith is/are:  
☐ Form PTO-1449 (PTO/SB/08A and 08B).  
☐ Copies of citations listed.  
b. ☒ will be transmitted within THREE MONTHS of the date of submission of requirements under 35 U.S.C. Sections 371(c).  
c. ☐ was previously submitted by applicant on \_\_\_\_\_  
Date
13. ☐ An assignment document is transmitted herewith for recording.
- A separate ☐ "COVER SHEET FOR ASSIGNMENT (DOCUMENT) ACCOMPANYING NEW PATENT APPLICATION" or ☐ FORM PTO 1595 is also attached.
- \_\_\_\_\_  
\_\_\_\_\_  
\_\_\_\_\_
14. ☒ Additional documents:  
a. ☒ Copy of request (PCT/RO/101)  
b. ☐ International Publication No. \_\_\_\_\_  
i. ☐ Specification, claims and drawing  
ii. ☐ Front page only  
c. ☐ Preliminary amendment (37 C.F.R. Section 1.121)  
d. ☐ Other
15. ☒ The above checked items are being transmitted  
a. ☒ before 30 months from any claimed priority date.  
b. ☐ after 30 months.
16. ☐ Certain requirements under 35 U.S.C. 371 were previously submitted by the applicant on \_\_\_\_\_, namely:  
\_\_\_\_\_  
\_\_\_\_\_  
\_\_\_\_\_

#### AUTHORIZATION TO CHARGE ADDITIONAL FEES

- ☒ The Commissioner is hereby authorized to charge the following additional fees that may be required by this paper and during the entire pendency of this application to Account No. 50-0850.

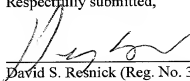
In re application of: ADDISON, P.S., et al. Group: Not yet assigned  
Application No.: Not Yet Assigned Examiner: Not yet assigned  
(National Phase Entry of PCT/GB00/01675)  
Filed: Herewith  
For: METHOD OF ANALYSIS OF MEDICAL SIGNALS

- ☒ 37 C.F.R. Section 1.492(a)(1), (2), (3), and (4) (filing fees)  
☒ 37 C.F.R. Section 1.492(b), (c) and (d) (presentation of extra claims)  
☒ 37 C.F.R. Section 1.17 (application processing fees)  
☐ 37 C.F.R. Section 1.17(a)(1)-(5)(extension fees pursuant to Section 1.136(a).  
☐ 37 C.F.R. Section 1.18 (issue fee at or before mailing of Notice of Allowance,  
pursuant to 37 C.F.R. Section 1.311(b))  
☐ 37 C.F.R. Section 1.492(e) and (f) (surcharge fees for filing the declaration  
and/or filing an English translation of an International Application later than 30  
months after the priority date).

Date: 01 November 2001

Respectfully submitted,

Customer No.: 26770

  
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1 "Method of Analysis of Medical Signals"

This invention relates to a method of analysis of medical signals, and in particular to a method of decomposition of cardiac signals using wavelet transform analysis. Specifically the invention relates to an improved method of resuscitation of patients in cardiac arrest.

10 In the UK, coronary heart disease is the second  
11 greatest contributor to deaths of people under 75. The  
12 social and economic consequences of these death rates

1 are enormous. The current survivability rates of  
2 patients after sudden cardiac failure are around 1:10.  
3  
4 Ventricular tachyarrhythmias, specifically ventricular  
5 fibrillation (VF), are the primary arrhythmic events in  
6 cases of sudden cardiac death. Administration of  
7 prompt therapy to a patient presenting with such  
8 symptoms can however lead to their successful  
9 resuscitation. Until recently, the only indicators of  
10 likelihood of survival of a patient to hospital  
11 discharge were traditional variables such as emergency  
12 service response time or bystander cardio-pulmonary  
13 resuscitation (CPR).  
14

15 In most cardiac complaints, analysis of a surface  
16 electrocardiogram (EKG) of the presenting patient is a  
17 rich source of information. However, until recently, a  
18 surface EKG recorded during VF and any subsequent  
19 medical intervention to defibrillate, was thought  
20 merely to present unstructured electrical activity, and  
21 not to provide useful information.  
22

23 The first attempts to derive prognostic information  
24 from EKGs of the heart in VF focussed on the importance  
25 of the amplitude of the waveform defined using peak-to-  
26 trough differences in the EKG voltage, measured as  
27 either the greatest deflection occurring in a  
28 predefined time slot, or as the average peak-to-trough  
29 voltage measured over a given time interval. It has  
30 been shown that the VF amplitude is inversely related  
31 to time elapsed since collapse, is a crude predictor of  
32 defibrillation outcome, and is a better indicator of



1 survival to hospital discharge than the traditional  
2 variables described above.

3

4 However, recording the VF amplitude accurately is  
5 significantly problematical. The EKG voltage amplitude  
6 measured during VF is dependent on the direction of the  
7 main fibrillation vector and is influenced by a variety  
8 of factors including patient chest shape; electrode  
9 size; electrode location; and skin/electrode interface  
10 resistance. This number of variables makes this  
11 amplitude measure both unreliable and inaccurate. That  
12 is, although the amplitude of the waveform of an EKG  
13 recorded during VF is now recognised to be a crude  
14 predictor of the likely outcome of resuscitation of a  
15 patient in VF, it is not a reproducible marker of  
16 sensitivity to defibrillation, and lacks clinical  
17 usefulness.

18

19 In a further development, it is also known to use Fast-  
20 Fourier based transforms to generate a frequency  
21 spectrum of an EKG in VF to analyse the signal. The  
22 median frequency (MF) divides the area under the  
23 spectrum into two equal parts. Since this plot is  
24 derived from information in both the voltage and time  
25 domains, external variables such as lead placement have  
26 less effect on the results than the method of observing  
27 the amplitude. However, CPR produces artefacts in the  
28 recorded EKG signal and, since pausing CPR merely to  
29 obtain an EKG signal free of artefacts is likely to  
30 compromise resuscitation, these artefacts are  
31 necessarily included in this frequency measure, and  
32 detract from its usefulness.

1

2 Thus the results of such signal analysis show some  
3 correlation with the likely outcome of resuscitation,  
4 but again lack sufficient sensitivity and specificity  
5 for clinical use. That is, this form of analysis has  
6 the disadvantage that, since the Fourier spectrum  
7 contains only globally averaged information, specific  
8 features in the signal are lost.

9

10 A method of accurate analysis of a surface EKG waveform  
11 recorded during VF would therefore be useful in  
12 understanding the pathophysiological processes in  
13 sudden cardiac death, and thus to produce a model for  
14 use:

15

16 in predicting the efficacy of therapy in individual  
17 cases; and

18

19 in determining the selection of the preferred course of  
20 primary, and alternative or adjunct therapies thus  
21 providing a means for individually tailored therapy for  
22 the specific patient needs

23

24 to improve the success rate of resuscitation of  
25 patients presenting in VF.

26

27 Atrial fibrillation (AF) is a common cardiac arrhythmia  
28 in older people. Atrial fibrillation can be stopped by  
29 giving an electric shock to the patient under general  
30 anaesthetic (cardioversion). However, many patient  
31 return to an AF rhythm soon after treatment. The  
32 technology detailed here may also provide a tool to

1 facilitate the clinical evaluation of AF exhibited in  
2 the electrocardiogram (EKG) so reducing the risk  
3 associated with general anaesthetic in patients where  
4 the applied therapy is likely to prove ineffective.

5

6 According to the present invention there is provided  
7 a method of decomposition of waveforms in a cardiac  
8 signal using wavelet transform analysis.

9

10 The method of the invention is non-invasive, accurate,  
11 and capable of delivering real-time information.

12

13 Preferably said method employs discretized wavelet  
14 transform analysis to process the EKG.

15

16 Preferably said method employs discretized continuous  
17 wavelet transform analysis to process the EKG.

18

19 Preferably said method comprises the steps of deriving  
20 the wavelet energy surfaces of an EKG signal; and  
21 plotting said wavelet energy surfaces against a  
22 location parameter  $b$ , and a scale parameter. The scale  
23 parameter may be dilation  $a$  or band pass frequency  $f_{bpc}$ .

24

25 The method initially comprises the steps of connecting  
26 electrodes to the presenting patient; and sampling the  
27 analogue input signal to derive the cardiac signal.

28

29 Typically said method comprises the step of visually  
30 displaying the cardiac signal.

31

1 Said method may display the distribution of energies  
2 within the cardiac signal. Said method may display  
3 coherent structures within the cardiac signal.

4

5 Said display may be by means of a contour plot. Said  
6 display may be by means of a surface plot. Preferably  
7 said method provides means to visualise the signal in  
8 real-time for clinical use.

9

10 Preferably said method is applicable in the analysis of  
11 an EKG in ventricular fibrillation.

12

13 Said method may be applicable in the analysis of an EKG  
14 in ventricular fibrillation after the commencement of  
15 cardio-pulmonary resuscitation (CPR).

16

17 The method may include the step of disassociating the  
18 component features of the temporal trace of a recorded  
19 EKG. Additionally or alternatively said method may  
20 include the step of temporal filtering of an EKG signal  
21 of a heart which is subject to CPR to disassociate the  
22 CPR signal from the heart signal.

23

24 Typically said method provides measurable  
25 characteristics for the estimation of the health of a  
26 heart in VF. Said method may provide measurable  
27 characteristics for the estimation of the health of a  
28 heart in AF. Said me may provide Typically said method  
29 provides measurable characteristics for the estimation  
30 of the health of a heart.

31

1 The method may provide measurable characteristics for  
2 the estimation of the time elapsed since the onset of a  
3 cardiac incident.

4  
5 Typically said method provides measurable  
6 characteristics for the estimation of the health of a  
7 heart after commencement in CPR.

8  
9 Said method may provide a prediction for the outcome of  
10 a given therapeutic intervention and so aid the  
11 clinical decision making process.

12  
13 Said method may provide a basis for individual, patient  
14 specific, protocols for therapeutic intervention.

15  
16 The method may provide a guide to the optimal timing of  
17 defibrillation of a heart in VF.

18  
19 Said method may include the step of constructing a  
20 damage index for reference purposes. Construction of  
21 said index might involve the development of a network  
22 classifier from a library of recorded data. Said  
23 network classifier may comprise a neural network. Said  
24 network classifier may comprise a wavelet network  
25 classifier.

26  
27 Application of the method of the invention represents a  
28 significant advance in coronary care by providing a  
29 reliable predictor of the outcome of shocking a patient  
30 in VF. In addition, the development of an algorithm  
31 using the method of the invention gives the ability to  
32 predict shock outcome and to facilitate individual

1 patient therapy. The ability to provide patient  
2 specific therapeutic intervention is a priority in the  
3 advancement of currently applied medical protocols.

4  
5 That is, as discussed above, in certain instances,  
6 after prolonged cardiac arrest preceding defibrillation  
7 pharmacological measures or CPR can increase the chance  
8 of successful resuscitation. Thus, employing the  
9 method to predict the outcome of shocking avoids futile  
10 defibrillation attempts which can even harm the heart,  
11 and can indicate the need for intervention, and  
12 influence the selection of the preferred type of  
13 intervention, to optimise the metabolic state of the  
14 heart prior to counter-shock.

15  
16 The predictor algorithm developed using the method is  
17 being tested using a new generation of defibrillation  
18 devices that have the flexibility to allow easy  
19 prototyping of the new defibrillation algorithms.

20  
21 According to a further aspect of the present invention  
22 there is provided a method of decomposition of  
23 waveforms in a cardiac signal using matching pursuit  
24 algorithms.

25  
26 According to a further aspect of the present invention  
27 there is provided an apparatus for decomposition of  
28 waveforms in a cardiac signal, said apparatus  
29 comprising wavelet transform analysis means.

30  
31 Said apparatus may include means to display the  
32 distribution of energies within a waveform.

1 Said apparatus may include a monitor adapted to display  
2 decomposed waveforms. Said apparatus may be adapted  
3 for inclusion in an EKG apparatus.

4  
5 According to a further aspect of the present invention  
6 there is provided defibrillation means adapted to  
7 operate in response to a signal generated by comparison  
8 of an EKG trace with decomposed waveform.

9  
10 That is, the invention preferably provides a method of  
11 wavelet analysis of cardiac signals which provides  
12 structural information about the heart - whether the  
13 heart is healthy or not - and has significant  
14 advantages over fast Fourier transforms.

15  
16 The invention may provide a display device in the form  
17 of a scrologram that provides real-time visualisation  
18 of a wavelet scalogram, showing the distribution of  
19 energies and coherent structures within the signal for  
20 use as guidance by a clinician.

21  
22 The invention may further provide a data analysis tool,  
23 which assists in shock timing (atrial pulsing). That  
24 is, the derived data may indicate the optimum time to  
25 administer shock to the heart. The invention may  
26 provide a damage index, preferably in the form of an  
27 artificial neural network.

28  
29 Preferably the invention provides dissociation of the  
30 component features of a temporal trace of a cardiac  
31 signal, which may for example be CPR, AF, or cardio-  
32 phonographic signals.

1    Embodiments of the invention will now be described by  
2    way of example only and with reference to the  
3    accompanying drawings in which:

4  
5  
6        Figure 1a is a Mexican hat wavelet;

7  
8        Figure 1b is the real part of a complex Morlet  
9    wavelet;

10  
11       Figure 2a is a schematic plot showing the dilation  
12    of a continuous wavelet;

13  
14       Figure 2b is a schematic plot showing the  
15    translation of a continuous wavelet;

16  
17       Figures 3a to Figure 3e are the plots of the  
18    'investigation' of a sinusoidal signal by Mexican  
19    hat wavelets of various sizes, showing the effect  
20    of translation of the wavelet along the signal  
21    (change in  $b$ ), and dilation of the wavelet (change  
22    in  $a$ );

23  
24       Figure 4a is the plot of five cycles of a sine  
25    wave of period  $P$ ;

26  
27       Figure 4b is the contour plot of  $T(a,b)$  against  $a$   
28    and  $b$  for the sine wave of Figure 4a;

29  
30       Figure 4c is the isometric surface plot of  $T(a,b)$   
31    against  $a$  and  $b$  for the sine wave of Figure 4a;

32



1 Figure 5a is the plot of a combination of two sine  
2 waves of period  $P_1$ , and  $P_2$ , where  $P_1 = 5P_2$ ;

3

4 Figure 5b is the contour plot of  $T(a,b)$  against  $a$   
5 and  $b$  for the sine wave of Figure 5a;

6

7 Figure 5c is the isometric surface plot of  $T(a,b)$   
8 against  $a$  and  $b$  for the sine wave of Figure 5a;

9

10 Figure 6a is an EKG trace of a pig heart in sinus  
11 rhythm;

12

13 Figure 6b is a 2D energy scalogram associated with  
14 the EKG trace of Figure 6a;

15

16 Figure 6c is a 3D energy scalogram associated with  
17 the EKG trace of Figure 6a;

18

19 Figures 6d, 6e, 6f and 6g are the energy surface  
20 plots from four segments of an EKG signal  
21 subsequent to the onset of VF, showing the three  
22 dominant ridges A, B, and C appearing in the  
23 transform surface, and showing in Figure 6g the  
24 onset of CPR after five minutes, associated with a  
25 gradual increase in passband frequency of the  
26 ridges A, B, and C;

27

28 Figure 7a is an energy scalogram for a pig heart  
29 for the first seven minutes of ventricular  
30 fibrillation, indicating the initiation of CPR  
31 after five minutes;

32

1 Figure 7b is a schematic diagram of the salient  
2 features of the scalogram of Figure 7a;  
3  
4 Figure 7c is the smoothed plot of energy at the  
5 8Hz level in the scalogram of Figure 7a against  
6 time;  
7  
8 Figure 8a is a typical segment of an EKG trace of  
9 a pig heart in VF;  
10  
11 Figures 8b, 8c, and 8d are the energy scalograms  
12 associated with the trace of Figure 8a;  
13  
14 Figure 9 is a screen shot of a real time viewer  
15 which shows the collected EKG data with its  
16 associated wavelet energy display in the form of  
17 its energy scalogram, where windows scroll to the  
18 right;  
19  
20 Figure 10a is a 7 second trace of human ECG  
21 showing a shock event;  
22  
23 Figure 10b is a scalogram corresponding to the  
24 trace of Figure 10a;  
25  
26 Figure 11a shows the proportion of energy in  
27 scalograms for 120 results (60 ROSC, and 60  
28 asystole) at 1.9 Hz after shocking;  
29  
30 Figure 11b shows the proportion of energy in  
31 scalograms for 120 results (60 ROSC, and 60  
32 asystole) at 9.3 Hz after shocking;

1

2 Figure 12a is a schematic representation of  
3 overlapping signal segments used in a neural  
4 network test study;

5

6 Figure 12b shows the weights attributed by the  
7 Kohonen network to the 30 frequency levels used in  
8 the scalogram;

9

10 Figure 13a is an aorta pressure trace;

11

12 Figure 13b shows the EKG for the same time period  
13 as the trace of Figure 13a; and

14

15 Figure 13c is the scalogram associated with the  
16 trace of Figure 13a derived from the Morlet  
17 wavelet;

18

19 Figure 13d is a detail of the phase part of  
20 scalogram Figure 13c;

21

22 Figure 13e is the scalogram associated with the  
23 trace of Figure 13a derived from the Mexican hat  
24 wavelet; and

25

26 Figure 13f demonstrates the correlation of aorta  
27 pressure pulse position with lines of zero phase;

28

29 Figures 14a is the plot of an EKG trace. Figure  
30 14b is its associated phase at around 1.5Hz.  
31 Figure 14c is its energy scalogram. The  
32 correlation of zero phase at this lower frequency

1 and high frequency (low dilation) peaks is thus  
2 illustrated.

3  
4 Figure 15a shows a 2 second segment of EKG taken  
5 from a patient with atrial fibrillation (AF).  
6 Figure 15b shows the wavelet scalogram plot  
7 associated with this EKG. Figure 15c shows the  
8 corresponding modulus maxima of the scalogram of  
9 Figure 15b.

10  
11 Figure 15d contains a 7 second segment of EKG  
12 exhibiting AF. Figure 15e is a trace of EKG  
13 temporal components with small amplitude. Figure  
14 15f shows the larger magnitude components i.e. the  
15 QRS and T waves.

16  
17 Figure 15g is a plot of a two second 'blow up' of  
18 part of the signal of Figure 15d; Figure 15h is a  
19 plot of a two second 'blow up' of part of the  
20 signal of Figure 15e; and Figure 15i is a plot of  
21 a two second 'blow up' of part of the signal of  
22 Figure 15f.

23  
24 Referring to the Figures, the present method employs  
25 the use of a wavelet transform to analyse a cardiac  
26 signal.

27  
28 The method involves the decomposition of the signal.  
29 This decomposition is accomplished by utilising wavelet  
30 transforms to decompose the signal in wavelet space.

31

1 A key distinction between the Fourier analysis of an  
2 EKG signal and its analysis by means of a wavelet  
3 function is that, whilst the Fourier transform employs  
4 a sinusoid function, a wavelet function is localised in  
5 time.

6  
7 The methodology for such decomposition may include  
8 discretized continuous wavelet transforms, orthonormal  
9 wavelet transforms of decimated construction, non-  
10 decimated wavelet transforms, wavelet packet transforms  
11 and matching pursuit algorithms.

12  
13 Signal processing employing wavelet transform analysis  
14 allows simultaneous elucidation of both spectral and  
15 temporal information carried within a signal. Such  
16 processing can employ either continuous or discrete  
17 transforms. The choice of wavelet transform used for a  
18 particular signal processing application depends on  
19 factors such as speed of computation necessary, the  
20 shape of signal specific features, the frequency  
21 resolution required, and the statistical analysis to be  
22 performed.

23  
24 The preferred method employs the discretized continuous  
25 transform as it provides high resolution in wavelet  
26 space at lower frequencies.

27  
28 This method thus employs the use of a discretized  
29 continuous wavelet transform to analyse a cardiac  
30 signal.

31

1 In particular, this method employs a wavelet transform  
2 as an interrogation tool for EKG signals of ventricular  
3 fibrillation.

4  
5 A variety of wavelet functions are available, and the  
6 most appropriate is selected to analyse the signal to  
7 be investigated.

8  
9 The wavelet transform of a continuous time signal,  
10  $x(t)$ , is defined as:

11  
12 
$$T(a,b) = \frac{1}{w(a)} \int_{-\infty}^{\infty} x(t) \overline{g\left(\frac{t-b}{a}\right)} dt$$
 equation 1

13

14 where  $g(t-b)/a$  is the analysing wavelet function and  
15  $\overline{\phantom{x}}$  denotes complex conjugate.  $w(a)$  is a scaling  
16 function usually of the form  $w(a)=a^n$  where  $n$  is usually  
17 1 or 0.5, and  $x(t)$ , in this application, is the single  
18 channel surface EKG time signal. The transform  
19 coefficients  $T(a,b)$  are found for both specific  
20 locations on the signal,  $b$ , and for specific wavelet  
21 dilations,  $a$ .  $T(a,b)$  is plotted against  $a$  and  $b$  in  
22 either a surface or contour plot.

23

24 While other wavelet types may be employed the wavelets  
25 mainly used in this method are: the Mexican hat wavelet  
26 and the Morlet wavelet, examples of which are shown in  
27 Figure 1.

1 The wavelet can translate along the signal (change in  
2  $b$ ) and dilate (change in  $a$ ). This is shown  
3 schematically in Figure 2 using a Mexican hat wavelet.

4 Figure 3 illustrates the way in which a sinusoidal  
5 signal can be 'investigated' at various locations by  
6 Mexican hat wavelets of various sizes. The numerical  
7 value of the convolution (equation 1) depends upon both  
8 the location and dilation of the wavelet with respect  
9 to the signal.

10 Figure 3a shows a wavelet of similar 'size' to the  
11 sinusoidal waves superimposed on the signal at a  $b$   
12 location which produces a reasonable matching of the  
13 wavelet and signal locally. From the Figure it is  
14 apparent that there is a high correlation between the  
15 signal and wavelet at this  $a$  and  $b$  location.  
16 Here, the cross correlation of the signal with the  
17 wavelet produces a large positive number  $T(a,b)$ .

18 Figures 3b and 3c show details of the wavelet transform  
19 of a signal using a wavelet of approximately the same  
20 shape and size as the signal in the vicinity of  $b$ .  
21 Figure 3b shows a wavelet of similar scale to the  
22 sinusoidal waveform located at maximum negative  
23 correlation. This produces a large negative  $T(a,b)$   
24 value. Figure 3c shows a wavelet of similar scale to  
25 the sinusoidal waveform located at a position on the  
26 time axis where near zero values of  $T(a,b)$  are  
27 realised. Figure 3d shows the effect on the transform  
28 of using the smaller  $a$  scale. It can be seen from the  
29 plot that the positive and negative parts of the  
30 wavelet are all in the vicinity of approximately the

1 same part of the signal, producing a value of  $T(a,b)$   
2 near zero. Figure 3e shows that the same thing happens  
3 when using a much larger wavelet, since the wavelet  
4 transform now covers various positive and negative  
5 repeating parts of the signal, again producing a near  
6 zero value of  $T(a,b)$ .

7

8 Wavelet transforms are not usually computed at  
9 arbitrary dilations for isolated locations in the  
10 signal, but rather over a range of  $a$  and  $b$ . A plot of  
11  $T(a,b)$  versus  $a$  and  $b$  for sinusoidal data using the  
12 Mexican hat wavelet is shown in Figure 4. Two methods  
13 are then employed to plot  $T(a,b)$ , namely a contour plot  
14 or scalogram as shown in Figure 4b, and a surface plot  
15 as shown in Figure 4c. At small and large values of  $a$ ,  
16 the near zero values of  $T(a,b)$  are evident from the  
17 plots, but at values of  $a$  of the order of one quarter  
18 of the wavelength of the sinusoid large undulations in  
19  $T(a,b)$  correlate with the sinusoidal forms of the  
20 signal.

21

22 Figure 5a shows two superpositioned sinusoidal  
23 waveforms, the first with period  $P1$ , the second with  
24 period  $P2$ .  $P1 = 5P2$ . Figures 5b and 5c, the transform  
25 plots of the superimposed waveforms clearly show the  
26 two periodic waveforms in the signal at scales of one  
27 quarter of each period. Thus, Figure 5 clearly  
28 demonstrates the ability of the continuous wavelet  
29 transform to decompose the signal into its separate



1 frequency components. That is, this transform  
2 'unfolds' the signal to show its constituent waveforms.  
3 The contribution to the signal energy at a specific  $a$   
4 scale and  $b$  location is proportional to the two-  
5 dimensional wavelet energy density function which is,  
6 in turn, proportional to the modulus of  $T(a,b)$ .

7

8 The method of the present invention thus involves the  
9 display of the transform as a contour plot. That is,  
10 the method is used to present information derived from  
11 an EKG trace of the heart in VF as a scalogram. The  
12 preferred form of presenting the information is as an  
13 energy scalogram, which presents the results as a plot  
14 showing the log of the wavelet energy coefficients,  
15 against the log of the bandpass centre frequency,  $f_{bpc}$ ,  
16 of the wavelets for each time increment. The bandpass  
17 centre frequency is proportional to the reciprocal of  
18 the dilation value,  $a$ . This plot highlights small  
19 changes in amplitude over the scales of interest. The  
20 transform copes with repeating features in time with  
21 shifting phase, making it appropriate for real time  
22 applications such as this.

23

24 That is, by performing continuous wavelet transform  
25 analysis on the ECG in VF, and then by producing an  
26 energy scalogram of the results, it is possible to  
27 unfold the signal in such a way that a previously  
28 hidden structure is apparent, in contrast to the  
29 apparently disorganised VF signal.

30

1 The method then includes quantifying the wavelet  
2 decomposition. This wavelet decomposition provides  
3 both qualitative visual and measurable features of the  
4 EKG in wavelet space.

5  
6 In practice, surface EKG tracings, recorded as soon as  
7 possible after the onset of VF, are analysed.

8  
9 As a demonstration of the efficacy of the method, in an  
10 example of an experimental procedure utilising this  
11 method of analysis employing wavelet techniques, VF was  
12 induced in anaesthetised pigs via a pacemaker probe,  
13 using a 90V impulse at 60 Hz. All of the pigs remained  
14 in VF, untreated for a period of either 3 or 5 minutes.  
15 After this time, CPR commenced. The surface EKG  
16 (standard lead II) was recorded using needle  
17 electrodes. The EKG was sampled at 300 Hz using a 12-  
18 bit A to D converter. The method of the present  
19 invention was then performed using 32 EKG tracings  
20 recorded immediately after the onset of VF.

21  
22 Figure 6a represents 4 beats of a pig heart in sinus  
23 rhythm. Figures 6b and 6c shows the wavelet transform  
24 of the signal displayed in two and three dimensions  
25 respectively.

26  
27 The QRS complex of the waveform is evident from the  
28 conical structures in Figure 6b converging to the high  
29 frequency components of the RS spike. The P and T  
30 waves are also labelled in the plot. The 3D landscape  
31 plot of Figure 6c shows the morphology of the signal in

1 wavelet space. In Figures 6b and 6c the continuous  
2 horizontal band (X) is associated with a frequency of  
3 1.7 Hz, the beat frequency of the sinus rhythm. The  
4 second band (Y) occurs at a frequency of approximately  
5 5.1 Hz, corresponding to the separation of the P-QRS-T  
6 components in time. At higher frequencies the P, QRS  
7 and T components are individually resolved according to  
8 their frequency makeup and temporal location.

9  
10 Figures 6d to 6g show the energy surfaces for four  
11 segments of EKG signal subsequent to the onset of VF,  
12 namely: (6d) 0-60 s; (6e) 60-100 s; (6f) 210-240 s;  
13 and (6g) 260-360 s.

14  
15 The morphology of the VF signal in wavelet space can be  
16 seen from the Figures to contain underlying features  
17 within a more complex surface topography. The most  
18 significant features are the dominant ridges that  
19 appear in the transform surface through time.

20  
21 Figure 6f shows these ridges quite clearly. A high-  
22 energy ridge can be observed at around 10 Hz and two  
23 lower energy bands can be observed at lower  
24 frequencies. These three ridges are labelled A, B and  
25 C, respectively, in the plot. Other ridges are also  
26 present within the scalogram.

27  
28 The energy surface in Figure 6g contains the onset of  
29 CPR after 5 min of untreated VF. The institution of  
30 CPR is associated with a gradual increase in the  
31 passband frequencies of ridges A, B and C. This change  
32 in the composition of the VF signal reflects electrical

1 changes in the fibrillating myocardium associated with  
2 the onset of CPR. This is because CPR produces  
3 antegrade myocardial blood flow and thus improves the  
4 metabolic state of the tissues, temporarily reversing  
5 the otherwise progressive decline in high band pass  
6 frequency components of the EKG wavelet decomposition.

7  
8 Figure 8a is a typical segment of an EKG trace of a pig  
9 heart in VF; Figures 8b, 8c, and 8d are the energy  
10 scalograms associated with the trace of Figure 8a. As  
11 clearly illustrated by these diagrams the principle  
12 dilation (band pass centre frequency) component of the  
13 scalogram is approximately 10Hz. However, using said  
14 method it is also apparent that this component is not  
15 constant. It 'pulses' with a degree of regularity. This  
16 structure is previously unreported.

17  
18 Figure 9 shows similar 'pulsing' in another porcine EKG  
19 signal. However, the structure is so pronounced that  
20 high energy, high frequency, intermittent components  
21 can be observed. These components have an occurrence  
22 frequency of the order of the original sinus rhythm:  
23 approximately 1.7Hz.

24  
25 Figure 10a is a human EKG signal segment containing a  
26 shock event. Figure 10b is the corresponding wavelet  
27 scalogram. It is apparent from the scalogram of Figure  
28 10b that both high frequency spiking and an  
29 intermittent high-energy region are present in the  
30 vicinity of 10 Hz and also above 10Hz.

31

1 The high frequency spiking is unique to the method of  
2 the present invention and is not visible using  
3 conventional Fourier techniques. The rich structure  
4 made visible within the EKG by the wavelet transform  
5 method is evident in the scalogram.

6 It is clearly seen from the Figures that applying the  
7 wavelet transform to an EKG signal of VF demonstrates  
8 that this signal is a rich source of valuable  
9 information. That is, it produces a display showing  
10 real time visualisation of the distribution of energies  
11 and coherent structures within the signal for use by a  
12 clinician in the selection of treatment strategies.

13 Using this method of analysis it is feasible to obtain  
14 real-time visual display of the EKG frequency  
15 characteristics in the wavelet domain during  
16 resuscitation. The scalogram produced provides  
17 information about the myocardium that is not available  
18 from a standard single channel surface EKG.

19

20 The wavelet scalogram decomposition can be displayed as  
21 a real time scrolling window, as shown in Figure 9.  
22 This window is useful as an aid for clinical decision  
23 making. It can be used as a stand-alone tool, or as  
24 basis for on-line statistical analysis of the current  
25 state of a heart.

26

27 To produce the window, a MATLAB TM R11 application is  
28 used. Each EKG sample taken results in the updating of  
29 a FIFO (First In First Out) buffer, and the EKG plot of  
30 Figure 9a. The scalogram of Figure 9b is then shifted

1 to the right and clipped before the 'missing' new right  
2 hand data is calculated, using conventional matrix  
3 algebra, and filled.

4  
5 This results in the two scrolling windows of Figure 9.  
6 The exponential ramp in the bottom right corner shows  
7 the compact support of the wavelet utilised at the  
8 given scale.

9  
10 Higher resolution scalograms are achieved through  
11 implementation on higher specification machines,  
12 purpose built hardware, or application specific  
13 software with coding using a lower level programming  
14 language, such as C++.

15  
16 CPR produces artefacts in the EKG signal. Additionally,  
17 this method delivers information the value of which is  
18 not degraded once the CPR artefacts are filtered from  
19 the EKG signal.

20  
21 From examination of the *scalograms* shown in Figures 6g,  
22 7a and 7b it can be seen that the VF signature and the  
23 signature of the CPR artefacts occupy distinct areas of  
24 the scalogram, which permits their separation.

25  
26 Known techniques such as the Modulus maxima method are  
27 now available to reduce the non-zero data points in the  
28 wavelet scalogram. This method reduces the topography  
29 of the scalogram surface to a series of ridges, thereby

1 considerably reducing the amount of data required to  
2 represent the signal in the wavelet space.

3

4 The modulus maxima obtained from a bandlimited signal  
5 with a wavelet of finite compact support in the  
6 frequency domain defines a complete and stable signal  
7 representation.

8

9 In this method, temporal filtering of the original EKG  
10 signal to disassociate the CPR signature from the heart  
11 signal can either be done directly, using the wavelet  
12 *energy scalograms*, or indirectly through modulus maxima  
13 techniques. This allows the heart to be monitored  
14 without necessitating cessation of CPR to allow rhythm  
15 recognition.

16

17 Further to the above, the method may also be applied to  
18 patients suffering from atrial fibrillation (AF) as a  
19 means of disassociating the prevalent QRS and T waves  
20 from the remainder of the signal.

21

22 Wavelet decomposition of the ECG signal is performed  
23 using an appropriate wavelet function. The modulus  
24 maxima technique is used to encapsulate the scalogram  
25 information in a series of ridges. Filtering of the  
26 signal is then undertaken using the modulus maxima  
27 information and through reconstruction the clinically  
28 useful information is isolated from the signal .

29

30 Specifically, Figure 15a shows the wavelet transform  
31 decomposition of a 2 second segment of ECG taken from a  
32 patient with atrial fibrillation. Below the ECG trace

1 is a wavelet scalogram plot. The corresponding modulus  
2 maxima of the scalogram is plotted below the scalogram.  
3  
4 For example, Figure 15d contains a 7 second segment of  
5 ECG exhibiting AF. The signal has been partitioned  
6 using a modulus maxima ridge following algorithm. The  
7 modulus maxima ridges have been separated into large  
8 and small scale features by thresholding the signal at  
9 a predetermined wavelet scale. A blow up of part of the  
10 signal is given in the lower three plots in the figure:  
11 Figures 15g, 15h and 15i. The middle of these plots  
12 contains the partitioned signal with the QRS complex  
13 and T wave filtered out revealing regular, coherent  
14 features that appear at a frequency of approximately  
15 400 beats per minute, typical of AF. The lower plot  
16 contains the partition with the filtered out QRS and T  
17 waves. Although, a relatively simple modulus maxima  
18 technique was used in this pilot study whereby the  
19 modulus maxima lines were simple partitioned into two  
20 subsets, the ability of the technique to separate the  
21 signal into QRS and T waves and underlying atrial  
22 activity is evident from the results. It is known that  
23 the decay in amplitude of a modulus maxima  
24 corresponding to a signal feature can be a function of  
25 the scale of the wavelet. It is possible to use this  
26 property to separate the ridge coefficients into a  
27 noisy and coherent part. In this way, further  
28 differentiation of the modulus maxima information can  
29 be implemented within a more sophisticated algorithm.  
30 This will facilitate the further separation of  
31 background noise, QRS and T waves, and atrial activity.  
32



1 This method thus facilitates useful interpretation of  
2 previously unintelligible EKG signals.

3 In patients presenting with uncoordinated rapid  
4 electric activity of the ventricle of heart, known as  
5 ventricular fibrillation (VF), there is no effective  
6 pulse and myocardial blood flow ceases. Even the  
7 institution of optimal cardio-pulmonary resuscitation  
8 (CPR) of the patient does not achieve more than 30% of  
9 the normal cardiac output. Ischaemia during cardiac  
10 arrest leads to a rapid depletion of myocardial high-  
11 energy phosphates, deterioration of transmembrane  
12 potentials, and disruption of intracellular calcium  
13 balance. Paradoxically, the myocardium in VF has  
14 supranormal metabolic demands. For this reason  
15 resuscitation attempts become less likely to succeed  
16 with the passage of time, and electrical defibrillating  
17 shocks increasingly result in asystole or EMD.

18

19 After prolonged cardiac arrest, the use of  
20 pharmacological measures or CPR before attempting  
21 defibrillation may increase the chances of successful  
22 resuscitation. This invention provides a robust and  
23 reliable method of analysis of the state of the  
24 myocardium in VF that prevents attempts to defibrillate  
25 at times that are unlikely to be successful, or even  
26 harmful to the heart. This method also provides an  
27 indication of the best way in which to optimise the  
28 metabolic state of the heart prior to counter-shock.

29

1 The method includes steps to establish a standard  
2 against which to evaluate collected data in a  
3 particular incidence.

4  
5 The method further employs use of measurable signal  
6 characteristics derived from the position and amplitude  
7 of features in the *scalogram* to estimate both the  
8 condition of the myocardium, and downtime of the  
9 subject while in VF.

10

11 The method thus provides for optimal treatment of the  
12 heart in VF, so fulfilling specific patient needs, by  
13 therapeutic intervention, if appropriate.

14

15 An energy scalogram such as that shown in Figure 7  
16 displays three distinct bands, labelled A, B, C. It is  
17 possible to derive quantifiable measures using  
18 correlations between the location and energy content of  
19 the bands.

20

21 Band A of Figure 7b represents the dominant energy band  
22 seen in the *scalogram* of Figure 7a, and corresponds to  
23 the tachycardic beating of VF. However the *scalogram*  
24 is much more informative in that it also shows, as  
25 bands B and C, the behaviour of other frequency  
26 components of the signal which were previously  
27 unreported.

28

29 Figure 7a shows a 2D energy scalogram. It includes the  
30 first 5 minute period of VF, followed by a 2.5 minute  
31 period of CPR. The onset of CPR is clearly identified  
32 by the distinct horizontal dark band in the lower right

1 quadrant of the Figure. Over the first 5 minute  
2 period, three bands, labelled A, B, C, can be clearly  
3 seen in the scalograms. These bands correspond to the  
4 ridges of Figures 6d to g. The increase in the  
5 frequency components of these three bands after the  
6 onset of CPR is evident in the plot. Bands B and C  
7 follow trajectories similar to each other in the  
8 *scalogram*, reducing in frequency over time. Band A,  
9 however, moves independently of the other two.  
10 Initially Band A increases, then it decreases to a  
11 local minimum value at approximately 70s. Between 70  
12 and 160s it increases relative to Bands B and C.  
13 Finally, it decreases until the start of CPR after  
14 300s. The same pattern was present in all 32 pig EKG  
15 traces of the experiment.  
16  
17 Obvious increases in the passband frequency of all  
18 three bands are observed in the *scalogram* after the  
19 onset of CPR. For some of the signals studied this  
20 increase in band C is masked by the dominant CPR band,  
21 and thus cannot be seen in the *scalogram*.  
22  
23 Figure 7b provides a schematic diagram of the salient  
24 features contained within the scalogram plots, where  $t_0$   
25 is immediately after the onset of VF;  $t_2$  is the start  
26 of CPR; and  $t_3$  is the end of the analysis. Figure 7c  
27 shows the relative proportion of energy contained in  
28 the scalogram in the 5 to 12 Hz region through time.  
29 There is an obvious decay in the relative energy  
30 associated with this region which is associated with  
31 the breakdown of co-ordinated activity in the heart.  
32

1 The steps of the method of the present invention  
2 described above establish that during the course of VF  
3 there is a reduction in the proportion of energy within  
4 the dominant frequency band indicated in Figure 7c.  
5 This dominant frequency band, Band A in Figure 7a, is  
6 demonstrated to be approximately 10 Hz for pig VF.

7

8 The energy within this band changes rapidly. This is  
9 illustrated by the 'pulses' in Figures 8,9,10.

10

11 The Figures 6,7,8,9,10 show that applying the wavelet  
12 transform to an EKG signal of VF demonstrates that this  
13 signal is a rich source of valuable information.

14

15 The underlying hypothesis of the method of the present  
16 invention is that the scalogram associated with an EKG  
17 correlates to the state of the myocardium as it decays  
18 subsequent to the onset of VF.

19

20 The method uses the information contained in the energy  
21 scalogram associated with an EKG to predict the likely  
22 success of clinical intervention, namely shocking.

23

24 It is therefore possible to develop a wavelet transform  
25 based tool for the prediction of shock outcome during  
26 ventricular fibrillation by:

27

- 28 1. collecting and collating data from sets of  
29 archived EKGs recorded from humans in VF where  
30 attempts to resuscitate by shocking were made; and  
31
- 32 2. developing a classifier for reference purposes.

1  
2 Figure 11 is a classification of the shock outcome in  
3 either asystole or a rhythmic response using a  
4 relatively simple statistical analysis. The experiment  
5 yielding the results to compile these Figures involved  
6 use of the lead II outputs of standard three lead EKGs  
7 of 120 patients in VF. Each trace is of three second  
8 duration sampled at 100 Hz. Of these patients, 60  
9 returned to sinus rhythm while the other 60  
10 deteriorated to asystole, post shock.

11  
12 Each trace was decomposed into an associated wavelet  
13 transform from which its energy scalogram was  
14 generated. The volume under this surface was then  
15 normalised to render the results independent of signal  
16 amplitude, but instead the result of the relative  
17 wavelet constituents of the signals. The log of the  
18 mean values at each dilation (band centre frequency)  
19 for each was then recorded. Figures 11a and 11b show  
20 the distribution of energies in a lower frequency band  
21 (1.9 Hz) and at the 9.3 Hz band. Clearly, through  
22 visual inspection, it is apparent that the proportion  
23 of energies around the 10 Hz band is higher for  
24 successful defibrillation attempts.

25  
26 The method then extends to apply neural techniques to  
27 analysis of wavelet pre-processed EKG signals.

28  
29 A pilot study conducted to determine the feasibility of  
30 using artificial neural techniques to provide a tool to  
31 predict the outcome of defibrillation during VF used  
32 eight human EKG trace segments containing shock events.

1 In these cases, the result of shocking was unequivocal  
2 - four patients returned to VF, and four experienced  
3 return of spontaneous circulation (ROSC).

4  
5 The traces were transformed using the Morlet wavelet,  
6 and energy scalograms containing thirty frequency  
7 levels were produced. This was then split into eight  
8 overlapping sections as shown in Figure 12a, each of  
9 200 points (2/3 seconds duration). These 200 location  
10 points were subsampled down to 50 to give eight  
11 scalograms for each trace of 50 x 30 elements. The  
12 volume under the energy scalograms were normalised and  
13 the patterns fed into a 'winner take all' Kohonen  
14 network with two output units and built in *conscience*  
15 (to avoid local minima). That is, the network was  
16 asked to group the 64 input patterns into two classes.  
17 All but ten outputs were collectively classified  
18 correctly giving a mean pattern error of 0.156 (against  
19 0.5 average pattern error expected from random inputs).

20  
21 Since this is a vector quantisation method (VQM) it was  
22 possible to identify how the network differentiates the  
23 patterns through inspection of its connective weights.  
24 The weights from each location position across all  
25 scales in the network are approximately the same, which  
26 means that there are no markers with which to  
27 synchronise the different pre-processed traces. This  
28 confirms that this neural network is too simple for  
29 this purpose. That is the network is not equipped to  
30 'consider' the relative phase of each input pattern.

31

1 Figure 12b shows the weights for the 'success' (ROSC)  
2 and 'failure' (VF) to the output units from the first  
3 two time slices across all scales. The weights  
4 indicate the classes are differentiated by the  
5 proportion of energy in the lower scales, which can be  
6 seen when compared with Figure 11.

7  
8 Although the above described method indicates the  
9 slight drop in the dominant frequency expected, the  
10 drop is very marginal which leads to the conclusion of  
11 the lack of competence of previously proposed methods  
12 as a defibrillation success predictor.

13  
14 In summary, a library of human ECG data containing data  
15 sets of human VF with attempts to resuscitate by  
16 shocking is used as a database. This database is  
17 extended to include data sets containing various  
18 methods of shocking including, for example, biphasic  
19 shocking. The biphasic shock waveform has resulted in  
20 an increased proportion of successful defibrillation  
21 attempts and is set to become the standard treatment  
22 for cases of VF.

23  
24 In one example, the recognised outcomes are defined by  
25 trace components of the post-shock window lasting until  
26 next shock (if present). If the ratio of the given  
27 rhythm exceeds 10% of the total window length the  
28 rhythms are prioritised according to the sequence:

29	30	31	32	Class	Rhythm	Ratio
	1			Pulse (SVR)	+10%	

1	2	No pulse (EMD)	+10%
2	3	Isoelectric (Asystole)	+10%
3	4	VF	+10%

4

5

6 Class 5 is the class of VF preceding shocks where VF  
7 re-establishes itself within 5 seconds following the  
8 shock (i.e. no change). The VF in all the other  
9 classes were non-VF in this period.

10

11 Wavelet analysis of this information in accordance with  
12 the method of the invention is then performed to:

13

14 construct a wavelet visualisation of the signal -  
15 usually by plotting wavelet energy surfaces against the  
16 location parameter  $b$  and the inverse of the dilation  
17 parameter  $a$ ;

18

19 provide measurable characteristics of the signal for  
20 estimation of downtime of the patient;

21

22 provide measurable characteristics of the signal for  
23 determining the health of the heart post CPR; and

24

25 to construct energy scalogram devised for the method -  
26 which uses the energy density function and the  
27 reciprocal of the wavelet a scale for use as a  
28 predictor tool.

29

30 As described above it is possible to use artificial  
31 neural network based techniques to develop such an  
32 indication of the state of myocardium. In the



1 alternative, it is possible to classify the wavelet  
2 scalogram through multilayered feedforward network  
3 types.

4  
5 The method may include the development of a modulus  
6 maxima algorithm tool for the preprocessing of ECG  
7 prior to its input into a neural network classifier.

8  
9 Using this technique improves network performance  
10 whether this data is further encoded, or presented as a  
11 whole, larger, sparse matrix as a pattern in the input  
12 space.

13  
14 This method therefore utilises the generalisation  
15 properties of a feed forward multi-layer network to  
16 predict the likelihood of defibrillation success from  
17 the wavelet transform of the EKG traces. This multi-  
18 layer network with its relatively simple dynamics, when  
19 combined with wavelet pre-processing, has proved itself  
20 a useful tool as a universal approximator.

21  
22 The classes of multi-layer network types of use in this  
23 method are:

- 24  
25 • Multi-layered feed forward (MLFF) neural networks  
26 with back propagation training and monotonic  
27 activation functions; and  
28 • Radial Basis Neural Networks (RBNN) as have  
29 previously been successfully applied to the denoising  
30 of medical Doppler ultrasound signals with wavelet  
31 preprocessing.

1  
2 As described above, the method involves the  
3 decomposition of EKG signals into a complete basis set  
4 defined by the wavelet shape and other parameters by  
5 salient basis functions of a different basis set,  
6 converged upon through regression techniques (sigmoid  
7 in the case of multilayer neural networks, Radial basis  
8 etc).

9  
10 These regression techniques can also be used to  
11 construct a wavelet basis function set directly.

12  
13 Methodologies for restricting the search space of the  
14 wavelet basis functions considered are known. Whilst  
15 this wavelet network has been shown to be effective for  
16 chaotic time series prediction, its implementation  
17 involves the use of wavelet *frames* of a decimated,  
18 dyadic, construction. The method of the present  
19 invention may employ continuous wavelet networks  
20 spanning a redundant wavelet *basis* which, although  
21 computationally more expensive, overcomes the time  
22 invariance constraint and the limited size of input  
23 space associated with use of wavelet frames.

24  
25 The method may use conventional gradient decent methods  
26 to produce a single layer wavelet classifier.

27  
28 These wavelet networks may be further employed as part  
29 of a multilayer system as a non-parameterised estimate  
30 of the original trace for input to further hidden  
31 layers.

32

1 The network type of choice for the automated prediction  
2 system of the method is selected on the basis of its  
3 sensitivity and selectivity in correctly classifying  
4 successful defibrillation outcomes in test set data,  
5 since this is most clinically useful.

6  
7 Thus experimental comparison of the three techniques  
8 demonstrates the efficacy of the wavelet transform  
9 technique.

10

11 The nature of underlying atrial activity can also be  
12 determined from wavelet decomposition of the EKG  
13 signal. The wavelet function gives information  
14 regarding the amplitude and, where appropriate, phase  
15 of the transformed signal. It is known that pressure  
16 readings taken from the aorta correlate to forms of  
17 atrial activity within the heart. Areas of localised  
18 high energy contained within the scalogram can be  
19 demonstrated to correlate with these pressure readings.  
20 This experimental result is extrapolated to mean that  
21 areas of localised high energy contained within the  
22 scalogram correlate with forms of atrial activity  
23 within the heart.

24

25 Figure 13a shows the aorta pressure, Figure 13b the EKG  
26 trace, for the same time period as Figure 13a, and  
27 Figure 13c shows the scalogram for the EKG of Figure  
28 13b. It is apparent that there is an increase in  
29 energy in the system during an atrial pulse, indicated  
30 by the dark blotches occurring in the scalogram at an  
31  $f_{bpc}$  of around 10 Hz. There is a frequency component  
32 between 1 and 2 Hz. As shown in Figure 13d, which

1 highlights the phase of the scalogram between 1 and 2  
2 Hz, it is apparent that the lines of zero phase are in  
3 alignment with the atrial pulse.  
4  
5 In a further scalogram, shown in Figure 13e, produced  
6 by using the Mexican hat wavelet transform which is  
7 real and has better temporal resolution, but worse  
8 frequency resolution than the complex scalogram of  
9 Figure 13c, it is demonstrated that positive high  
10 amplitude components are shown at the same positions  
11 for scales of between 1 and 2 Hz, thus reinforcing the  
12 findings extrapolated from Figure 13c. That is as  
13 shown in Figure 13f, the lines of zero phase correlate  
14 with the pulse position.  
15  
16 The lines of zero phase within the 1.8Hz frequency band  
17 also align with regular peaks in the scalograms, as  
18 shown in Figures 14a, 14b & 14c. This links the  
19 presence of the 1.8 Hz band with the observed peaks at  
20 higher frequencies. This correlation between the 1.8  
21 Hz band and the aorta pressure pulse suggests atrial  
22 activity is present.  
23  
24 In a further application of the method, means for  
25 identifying the optimum timing for application of the  
26 defibrillation shock can be extrapolated from the  
27 pulsing identified by the wavelet technique and shown  
28 in Figures 8, 9, 10, and 14, by comparison with traces  
29 of attempts at defibrillation which initially fail but  
30 are subsequently successful.

1  
2 Thus, any data sets, in the above, that correspond to  
3 multiple shocking of the same patient, where  
4 defibrillation has been repeatedly attempted are  
5 considered separately since these traces hold important  
6 information.

7  
8 The pilot study detailed above used Morlet wavelet  
9 based energy scalogram decomposition of signal segments  
10 immediately prior to shocking. A full parametric  
11 wavelet study of the method determines the optimum  
12 method.

13  
14 The method includes the development of a classifier  
15 using the wavelet transform analysis.

16  
17 Various types of neural network classifier are  
18 achievable using this method.

19  
20 The linkage of shock timing to the phase information of  
21 wavelet components allows for increased defibrillation  
22 success and reduced shock energies. The wavelet-  
23 derived information can also be employed to predict the  
24 likelihood of shock success, preventing futile or  
25 harmful defibrillation attempts, and providing a  
26 predictor of an optimal resuscitation strategy or  
27 strategies.

28  
29 This method demonstrates the utility of the wavelet  
30 transform as a new method of EKG signal analysis during  
31 VF. It provides a robust, real-time solution to the

1 problem of useful monitoring of the myocardium during  
2 resuscitation.

3

4 When compared with conventional statistical methods,  
5 such as fast Fourier transforms, it is seen that the  
6 temporal resolution of the wavelet technique gives a  
7 scalogram which better describes the non-stationary,  
8 intermittent, nature of the EKG trace to be analysed,  
9 and gives a method of greater predictive effectiveness  
10 than is already known. The effectiveness criteria for  
11 the networks of the method of the present invention are  
12 based upon their sensitivity and selectivity in  
13 correctly classifying successful defibrillation  
14 outcomes from test data sets.

15

16 Although this description refers to wavelet transform  
17 analysis, this term is to be construed to include  
18 matching pursuit algorithms and similar analysis  
19 techniques.

20

21 Modifications and improvements can be made to the above  
22 without departing from the scope of the invention.

## 1 CLAIMS

2

3 1. A method of decomposition of waveforms in a  
4 cardiac signal using wavelet transform analysis.

5

6 2. A method as claimed in Claim 1 comprising the step  
7 of employing discretized wavelet transform  
8 analysis to process the said waveform.

9

10 3. A method as claimed in Claim 1 comprising the step  
11 of employing discretized continuous wavelet  
12 transform analysis to process the cardiac  
13 waveform.

14

15 4. A method as claimed in any preceding claim  
16 comprising the steps of deriving the wavelet  
17 energy surfaces of an electrocardiogram (EKG)  
18 signal; and plotting said wavelet energy surfaces  
19 against a location parameter  $b$ , and a scale  
20 parameter.

21

22 5. A method as claimed in Claim 4 wherein said scale  
23 parameter is dilation  $a$ .

24

25 6. A method as claimed in Claim 4 wherein said scale  
26 parameter is band pass frequency  $f_{bpc}$ .

27

28 7. A method as claimed in any preceding claim  
29 comprising the initial steps of connecting  
30 electrodes to a presenting patient; and sampling  
31 the analogue input signals recorded to derive the  
32 cardiac signal.

- 1 8. A method as claimed in any preceding claim  
2 including visually displaying the cardiac signal.  
3
- 4 9. A method as claimed in any preceding claim  
5 including visually displaying the distribution of  
6 energies within the cardiac signal.  
7
- 8 10. A method as claimed in any preceding claim  
9 including visually displaying coherent structures  
10 within the cardiac signal.  
11
- 12 11. A method as claimed in any preceding claim  
13 including visually displaying the signal in real-  
14 time for clinical use.  
15
- 16 12. A method as claimed in any preceding claim  
17 comprising the step of constructing a contour plot  
18 to display the decomposed waveform obtained.  
19
- 20 13. A method as claimed in any preceding claim  
21 comprising the step of constructing a surface plot  
22 to display the decomposed waveform obtained.  
23
- 24 14. A method as claimed in any preceding claim  
25 comprising the step of constructing a 2D or a 3D  
26 energy scalogram to display the decomposed  
27 waveform obtained.  
28
- 29 15. A method as claimed in any preceding claim  
30 including the step of disassociating the component  
31 features of the temporal trace of a recorded EKG.  
32



- 1 16. A method for the analysis of an EKG of a heart in  
2 ventricular fibrillation including the method as  
3 claimed in any preceding claim.  
4
- 5 17. A method for the analysis of an EKG of a heart in  
6 ventricular fibrillation after the commencement of  
7 cardio-pulmonary resuscitation (CPR) including the  
8 method as claimed in any of Claims 1 to 15.  
9
- 10 18. A method as claimed in Claim 17 including the step  
11 of temporal filtering of the EKG signal of a heart  
12 that is subject to CPR to disassociate the CPR  
13 signal from the heart signal.  
14
- 15 19. A method as claimed in Claim 17 or Claim 18 using  
16 *wavelet energy scalograms*.  
17
- 18 20. A method as claimed in Claim 17 or Claim 18 using  
19 ridge following techniques  
20
- 21 21. A method as claimed in Claim 20 wherein said ridge  
22 following techniques are modulus maxima  
23 techniques.  
24
- 25 22. A method for the estimation of the health of a  
26 heart in VF including the method of any of Claims  
27 1 to 15 to provide measurable characteristics.  
28
- 29 23. A method as claimed in Claim 22 wherein said  
30 measurable characteristics are used to provide an  
31 estimate of the time elapsed since the onset of a  
32 cardiac incident.

- 1 24. A method as claimed in Claim 22 wherein said  
2 measurable characteristics are used to provide an  
3 estimate of the health of a heart after  
4 commencement of CPR.  
5
- 6 25. A method as claimed in any of Claims 22 to 24  
7 wherein said measurable characteristics are used  
8 to predict the outcome of a given therapeutic  
9 intervention.  
10
- 11 26. A method as claimed in any of Claims 22 to 25  
12 wherein said measurable characteristics are used  
13 to provide a guide for the optimal timing of  
14 defibrillation of a heart in VF.  
15
- 16 27. A method for the analysis of an EKG of a heart in  
17 atrial fibrillation including the method as  
18 claimed in any of Claims 1 to 14.  
19
- 20 28. A method as claimed in Claim 27 including the step  
21 of partitioning the signal to provide separate  
22 traces of QRS and T waves, and/or atrial activity  
23 and/or background noise.  
24
- 25 29. A method as claimed in any preceding claim  
26 including the step of constructing a damage index  
27 for reference purposes.  
28
- 29 30. A method as claimed in Claim 29 wherein  
30 construction of said index includes the step of  
31 developing network classifier from a library of  
32 recorded data.

- 1 31. A method as claimed in Claim 30 wherein said  
2 network classifier developed is a neural network.  
3
- 4 32. A method as claimed in any of Claims 29 to 31  
5 wherein said network classifier developed is a  
6 wavelet network classifier.  
7
- 8 33. A method of decomposition of cardiac waveforms  
9 using matching pursuit algorithms.  
10
- 11 34. Apparatus for decomposition of waveforms in a  
12 cardiac signal, said apparatus comprising wavelet  
13 transform analysis means.  
14
- 15 35. Apparatus as claimed in Claim 34 including means  
16 to display the distribution of energies within a  
17 waveform.  
18
- 19 36. Apparatus as claimed in Claim 34 or Claim 35  
20 including a monitor adapted to display decomposed  
21 waveforms.  
22
- 23 37. Apparatus as claimed any of Claims 34 to 36  
24 adapted for inclusion in an EKG apparatus.  
25
- 26 38. Defibrillation means adapted to operate in  
27 response to a signal generated by comparison of an  
28 EKG trace with decomposed waveform obtained by the  
29 method of any of Claims 1 to 33.  
30

1 39. A method as described in any of Claims 1 to 33  
2 with reference to or as shown in the accompanying  
3 drawings.

4  
5 40. Apparatus as described in any of Claims 34 to 38  
6 with reference to or as shown in the accompanying  
7 drawings.

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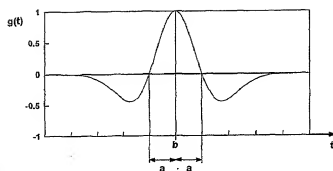


Figure 1(a)

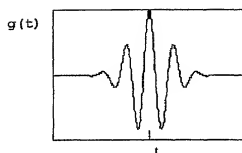


Figure 1(b)

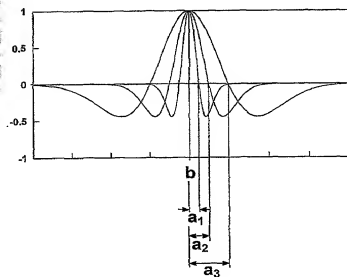


Figure 2(a)

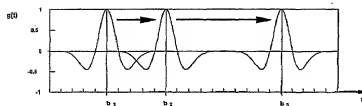
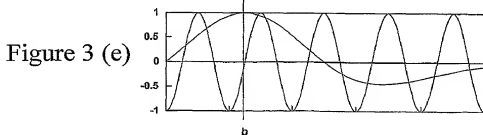
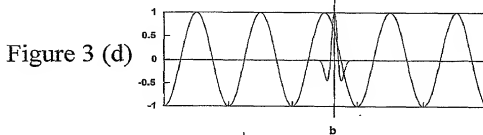
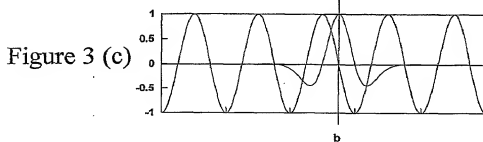
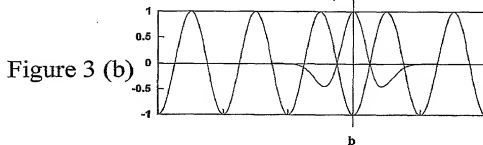
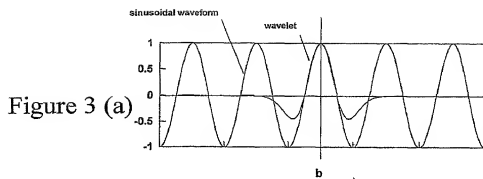


Figure 2(b)

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Figure 4 (a)

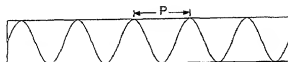


Figure 5 (a)

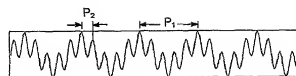


Figure 4 (b)

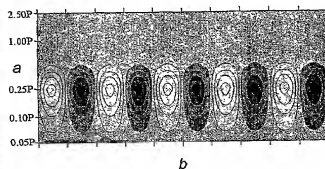


Figure 5 (b)

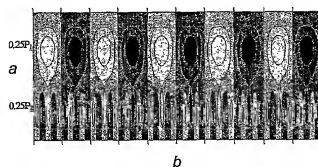


Figure 4 (c)

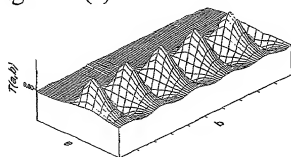
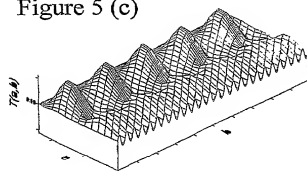


Figure 5 (c)



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Figure 6 (a)



Figure 6 (b)

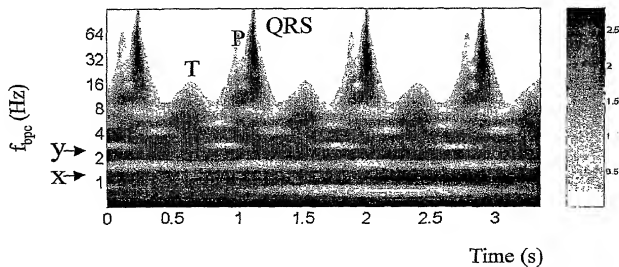
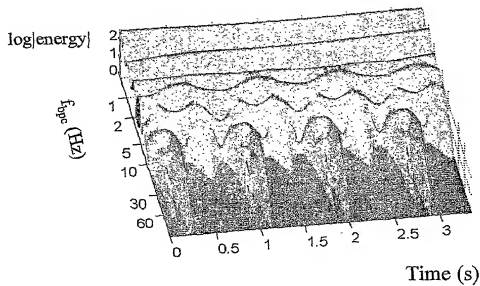


Figure 6 (c)





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Figure 6 (d)

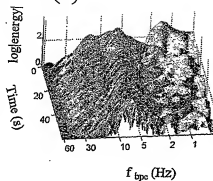


Figure 6 (e)

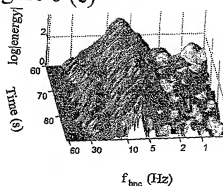


Figure 6 (f)

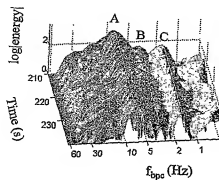
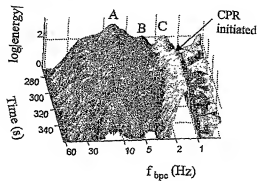


Figure 6 (g)



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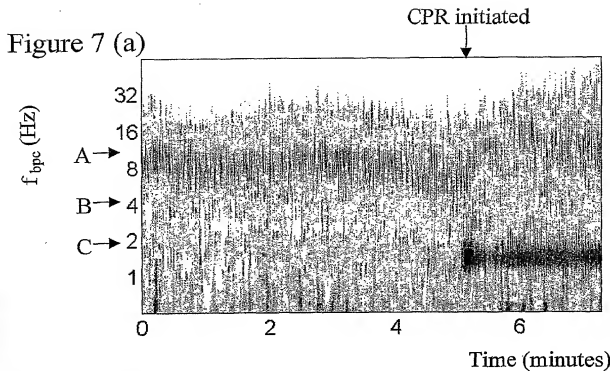


Figure 7 (b)

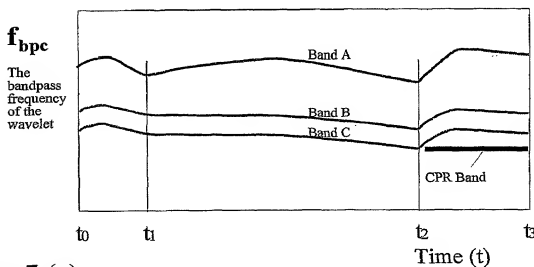
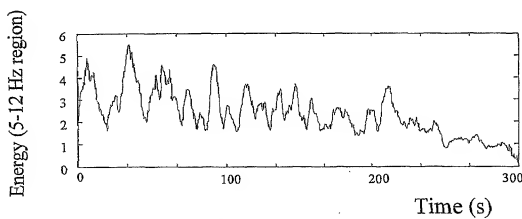


Figure 7 (c)



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Figure 8 (a)

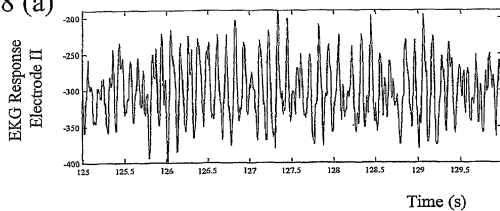


Figure 8 (b)

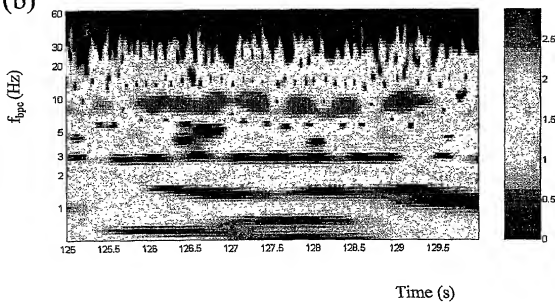


Figure 8 (c)

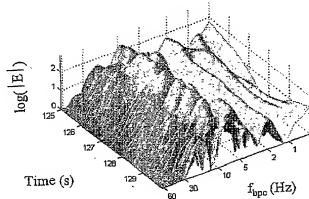
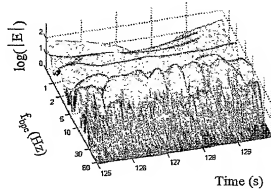


Figure 8 (d)



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Figure 9

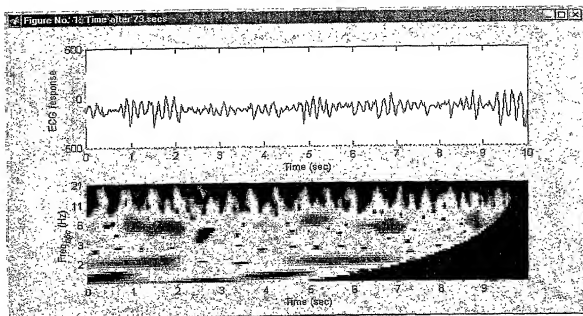


Figure 10 (a)

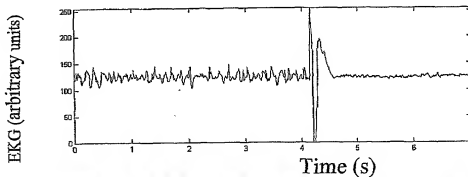
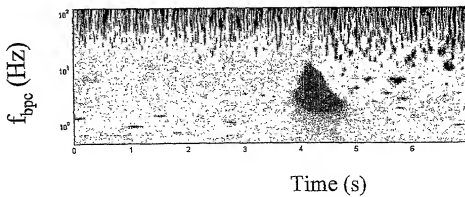


Figure 10 (b)



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Figure 11 (a)

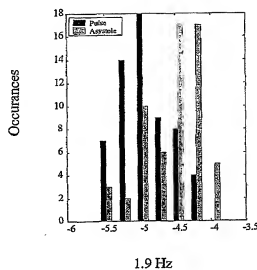


Figure 11 (b)

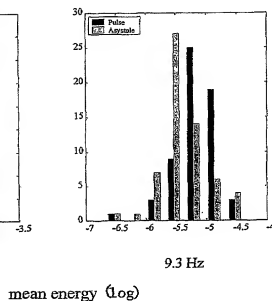


Figure 12 (a)

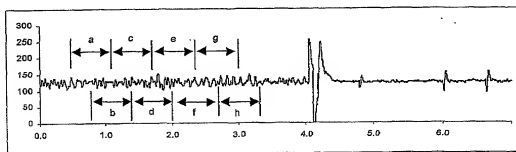
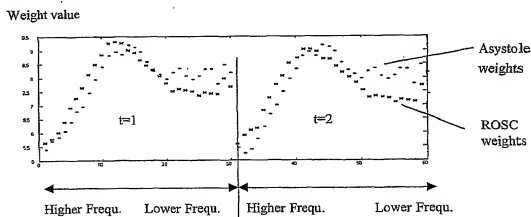


Figure 12 (b)



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Figure 13 (a)

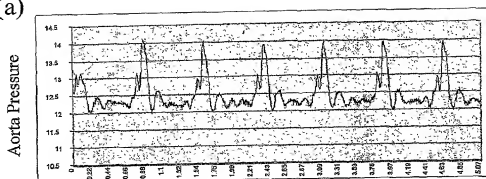


Figure 13 (b)

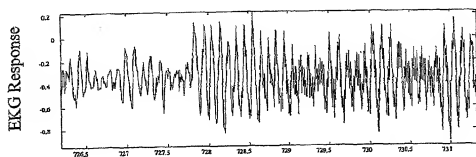
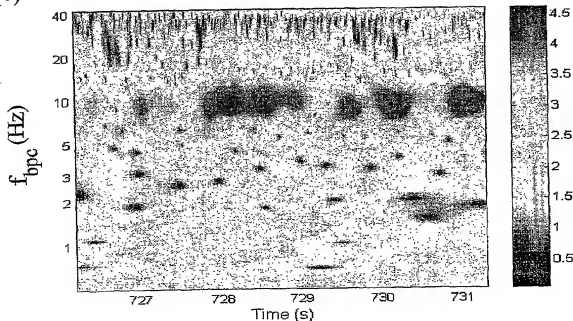


Figure 13 (c)



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Figure 13 (d)

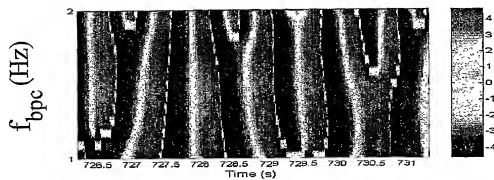


Figure 13 (e)

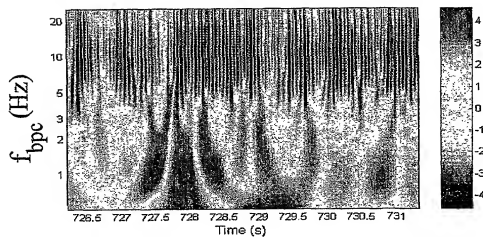
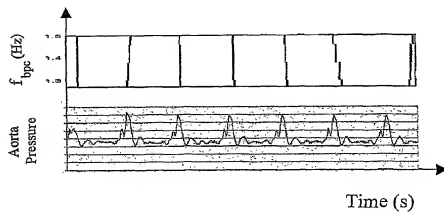


Figure 13 (f)



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Figure 14 (a)

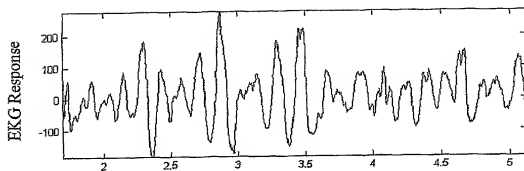


Figure 14 (b)

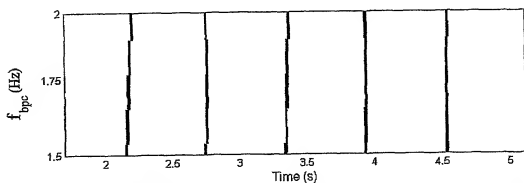
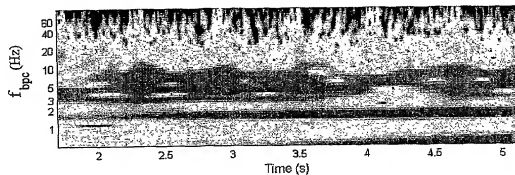


Figure 14 (c)





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Figure 15 (a)

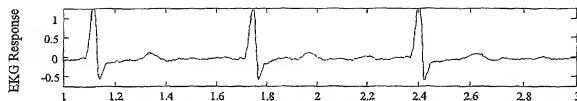


Figure 15 (b)

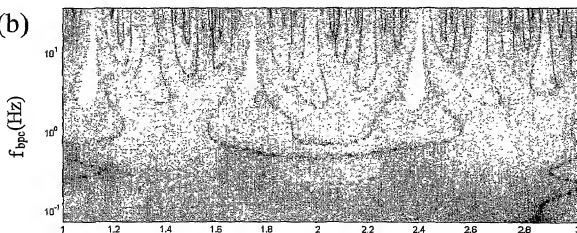
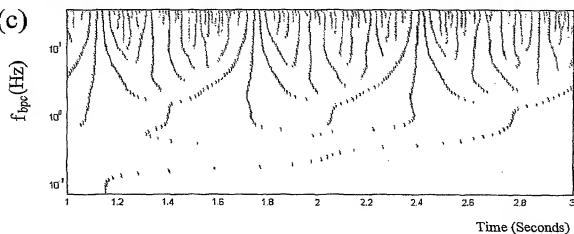


Figure 15 (c)



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Figure 15 (d)

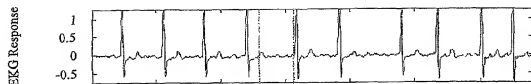


Figure 15 (e)

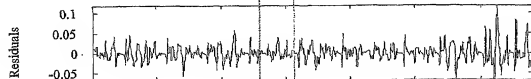


Figure 15 (f)

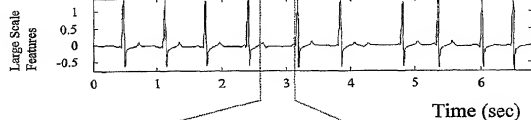


Figure 15 (g)

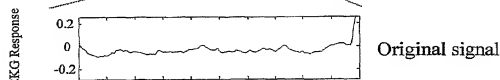


Figure 15 (h)

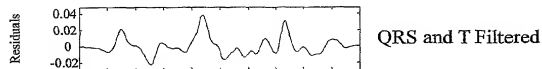
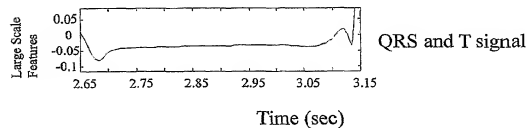


Figure 15 (i)



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### DECLARATION AND POWER OF ATTORNEY

As a below named inventor, I hereby declare that: My residence, post office address and citizenship are as stated below next to my name. I believe I am the original, first and sole inventor (if only one name is listed at 201) below or an original, first and joint inventor (if plural names are listed at 201-208 below) of the subject matter which is claimed and for which a patent is sought on the invention entitled:

"Method of Analysis of Medical Signals"

which is described and claimed in:

- ☐ the specification attached hereto.
- ☒ the specification in PCT Application Serial Number PCT/GB00/01675 ✓  
filed on 2 May 2000 ✓

I hereby state that I have reviewed and understand the contents of the above identified specification, including the claims, as amended by any amendment referred to above. I acknowledge the duty to disclose information which is material to the examination of this application in accordance with Title 37, Code of Federal Regulations, §1.56(a). I hereby claim foreign priority benefits under Title 35, United States Code, §119 of any foreign application(s) for patent or inventor's certificate listed below and have also identified below any foreign application for patent or inventor's certificate having a filing date before that of the application on which priority is claimed.

Prior Foreign/PCT Applications and Any Priority Claims Under 35 U.S.C. §119:			
Application No.	Filing Date	Country	Priority Claimed under 35 U.S.C. §119?
9910019.0 ✓	1 May 1999 ✓	United Kingdom ✓	<input checked="" type="checkbox"/> YES <input type="checkbox"/> NO
9916499.8 ✓	15 July 1999 ✓	United Kingdom ✓	<input checked="" type="checkbox"/> YES <input type="checkbox"/> NO
9919677.6 ✓	20 August 1999 ✓	United Kingdom ✓	<input checked="" type="checkbox"/> YES <input type="checkbox"/> NO
9923110.2 ✓	1 October 1999 ✓	United Kingdom ✓	<input checked="" type="checkbox"/> YES <input type="checkbox"/> NO
0003711.9 ✓	17 February 2000 ✓	United Kingdom ✓	<input checked="" type="checkbox"/> YES <input type="checkbox"/> NO

I hereby claim the benefit under 35 U.S.C. §120 of any United States application(s) or PCT international application(s) designating the United States of America that is/are listed below, and insofar as the subject matter of each of the claims of this application is not disclosed in that/those prior application(s) in the manner provided by the first paragraph of 35 U.S.C. §112, I acknowledge the duty to disclose material information as defined in 37 CFR §1.56(a) which occurred between the filing date of the prior application(s) and the national or PCT international filing date of this application.

Prior U.S. Applications or PCT International Applications Designating the U.S.-Benefit under 35 U.S.C. §120				
U.S. Applications		Status (Check One)		
Application Serial No.	U.S. Filing Date	Patented	Pending	Abandoned

PCT Applications Designating the U.S.				
Application No.	Filing Date	U.S. Serial No. Assigned		

**CLAIM FOR BENEFIT OF PRIOR U.S. PROVISIONAL APPLICATION(S)  
(35 U.S.C. §119(e))**

I hereby claim the benefit under Title 35, United States Code, §119(e) of any United States provisional application(s) listed below:

Applicant	Provisional Application Number	Filing Date

**POWER OF ATTORNEY** As a named inventor, I hereby appoint the following attorney(s) with full powers of association, substitution and revocation to prosecute this application and transact all business in the Patent and Trademark Office connected therewith.

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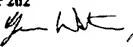
2 0 1	FULL NAME OF INVENTOR	LAST NAME <u>ADDISON</u>	FIRST NAME <u>PAUL</u>	MIDDLE NAME <u>STANLEY</u>
	RESIDENCE & CITIZENSHIP	CITY	STATE OR FOREIGN COUNTRY <u>UNITED KINGDOM</u>	COUNTRY OF CITIZENSHIP <u>A BRITISH SUBJECT</u> ✓
	POST OFFICE ADDRESS	POST OFFICE ADDRESS <u>68 BUCKSTONE CROOK</u>	CITY <u>EDINBURGH</u> <i>GBX</i>	STATE OR COUNTRY AND ZIP CODE <u>UNITED KINGDOM EH10 6UR</u>

0900070-1111111

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	RESIDENCE & CITIZENSHIP	CITY	STATE OR FOREIGN COUNTRY <u>UNITED KINGDOM</u>	COUNTRY OF CITIZENSHIP <u>A BRITISH SUBJECT</u> ✓
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2 0 3	FULL NAME OF INVENTOR	LAST NAME	FIRST NAME	MIDDLE NAME
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	POST OFFICE ADDRESS	POST OFFICE ADDRESS	CITY	STATE OR COUNTRY AND ZIP CODE

I hereby further declare that all statements made herein of my own knowledge are true and that all statements made on information and belief are believed to be true; and further, that these statements were made with the knowledge that wilful false statements and the like so made are punishable by fine or imprisonment, or both, under Section 1001 of Title 18 of the United States Code, and that such wilful false statements may jeopardize the validity of the application or any patent issued thereon.

Signature of Inventor 201 	Date: 30 OCT 2001
Signature of Inventor 202 	Date: 30 - 10 - 2001
Signature of Inventor 203	Date:

0998070-14101